

1 **Vegetation structure and fire weather influence variation in**
2 **burn severity and fuel consumption during peatland**
3 **wildfires**

4

5 **G. Matt Davies^{1,2}, Rut Domènech^{3,4}, Alan Gray⁵ and Paul C. D. Johnson²**

6 [1]{School of Environment and Natural Resources, Kottman Hall, The Ohio State University,
7 Columbus, Ohio, 43210, USA }

8 [2]{Boyd Orr Centre for Population and Ecosystem Health, Institute for Biodiversity, Animal
9 Health and Comparative Medicine, University of Glasgow, Graham Kerr Building, Glasgow,
10 G12 8QQ, United Kingdom}

11 [3]{Solway Centre for Environment and Culture, University of Glasgow, Henry Duncan
12 Building, Crichton University Campus, Dumfries, DG1 4ZL, United Kingdom}

13 [4]{Forest Sciences Centre of Catalonia (CTFC), Ctra. de Sant Llorenç de Morunys, Km 2,
14 25280 Solsona, Spain}

15 [5]{Centre for Ecology and Hydrology, Bush Estate, Penicuik, Midlothian, EH26 0QB,
16 United Kingdom}

17 Correspondence to: G. M. Davies (davies.411@osu.edu)

18

19 **Abstract**

20 Temperate peatland wildfires are of significant environmental concern but information on
21 their environmental effects is lacking. We assessed variation in burn severity and fuel
22 consumption within and between wildfires that burnt British moorlands in 2011 and 2012. We
23 adapted the Composite Burn Index (pCBI) to provide semi-quantitative estimates of burn
24 severity. Pre- and post-fire surface (shrubs and graminoids) and ground (litter, moss, duff)
25 fuel loads associated with large wildfires were assessed using destructive sampling and

1 analysed using a Generalised Linear Mixed Model (GLMM). Consumption during wildfires
2 was compared with published estimates of consumption during prescribed burns. Burn
3 severity and fuel consumption were related to fire weather, assessed using the Canadian Fire
4 Weather Index System (FWI System), and pre-fire fuel structure. pCBI varied 1.6 fold
5 between, and up to 1.7 fold within, wildfires. pCBI was higher where moisture codes of the
6 FWI System indicated drier fuels. Spatial variation in pre- and post-fire fuel load accounted
7 for a substantial proportion of the variance in fuel loads. Average surface fuel consumption
8 was a linear function of pre-fire fuel load. Average ground fuel combustion completeness
9 could be predicted by the Buildup Index. Carbon release ranged between 0.36 kg C m^{-2} and
10 1.00 kg C m^{-2} . The flammability of ground fuel layers may explain the higher C release-rates
11 seen for wildfires in comparison to prescribed burns. Drier moorland community types appear
12 to be at greater risk of severe burns than blanket-bog communities.

13 **1 Introduction**

14 Peatland wildfires pose a significant global challenge due to their potential for severe effects
15 on ecosystem functioning and the detrimental role they may play in climate change. Peatlands
16 account for approximately 2.5 % of Earth's land-cover (Kaat and Joosten 2009) and contain
17 more than 600 Gt of stored carbon (Yu et al. 2010), equivalent to 25 % of global soil organic
18 carbon stocks (Mitra et al. 2005) and 75 % of all atmospheric carbon (Kaat and Joosten 2009).
19 The degradation of this resource is a potential positive feedback to climate change and
20 smouldering wildfires also have other significant environmental and human impacts such as
21 respiratory problems associated with the inhalation of noxious smoke, the significant effort
22 and costs involved in fire fighting, destruction of soil seedbanks, widespread plant mortality
23 and post-fire erosion and water pollution problems (Watts and Kobziar 2013). Increased fire
24 risk and severity with climate change means wildfires pose a particularly significant threat to
25 the ecological integrity and carbon stocks of peatlands (Turetsky et al. 2015).

26 The majority of research on the effects of peatland wildfires has come from tundra, boreal and
27 tropical ecosystems (Turetsky et al. 2015). Temperate peatlands are also an important carbon
28 store and habitat type but many have a long history of disturbance and management (e.g.
29 Moore 2002). British, peatlands are acknowledged to be of significant national and
30 international conservation importance though most have been subjected to a variety of land
31 management practices, including burning and grazing, over at least the last two centuries
32 (Bonn et al. 2009). Many peatlands have also been significantly impacted by drainage

1 (Holden et al. 2004) and nutrient deposition from atmospheric pollution (Hogg et al. 1995).
2 British peatland habitats contain fire-prone vegetation including moorlands dominated by
3 *Calluna vulgaris* L. Hull (hereafter *Calluna*) and a variety of mire and bog communities
4 associated with *Molinia caerulea* (L.) Moench and *Eriophorum* spp. The majority of such
5 habitats are underlain either by deep peat deposits or by shallower organic soils that
6 nevertheless hold substantial amounts of carbon. Estimates suggest that around 88 t C ha⁻¹ are
7 stored in the soil and up to 2 t C ha⁻¹ in the vegetation of dwarf shrub dominated moorlands in
8 the UK (Ostle et al. 2009). The majority of the U.K.'s 4.5 Tg of soil carbon stocks are stored
9 in peat deposits below heath, bog and moorland habitats (Bradley et al. 2005). Managed
10 burning is an important control on the structure of these habitats with fires burnt regularly in
11 both moorland and blanket bog habitats systems (Bonn et al. 2009). Recommended burn
12 rotations are 15-25 years for *Calluna*-dominated moorlands whilst longer rotations or no
13 burning are recommended for wetter bog communities (Scottish Government 2011). The role
14 of fire in peatland ecology has become a highly controversial subject with substantial debate
15 surrounding the effect of managed burning on ecosystem dynamics (e.g. Grant et al. 2012).
16 The situation is not helped by a lack of data on how fire affects temperate peatland
17 ecosystems such as those found in the UK. A number of studies have been completed but
18 mostly for low severity experimental prescribed burns (e.g. Davies et al. 2010) or, for a few
19 individual wildfire events (e.g. Davies et al. 2013, Maltby et al. 1990, Worrall et al. 2011).
20 There is a consensus that wildfires pose a substantial and growing threat in the context of a
21 changing climate (Bonn et al. 2009). In this context, data is urgently-needed on both the scale
22 of the wildfire problem and the effects of such burns.

23 In systems with peat or organic soils severe wildfires that ignite carbon-rich deposits can lead
24 to substantial, instantaneous losses of carbon (Davies et al. 2013) and long-term changes to
25 ecosystem function (Maltby et al. 1990). Whilst the severe effects of smouldering peat fires
26 are obvious, such burns lie at one end of a spectrum of burn severity and not all fires on
27 peatlands necessarily ignite peat or cause ecological damage. Indeed carefully managed
28 burning of peatlands can have a variety of ecosystem benefits (Davies et al. 2008a).
29 Differences in burn severity can be caused by between and within site variation in fuel type
30 and fuel structure as well as by differences in fire weather conditions (e.g. fuel moisture
31 content, wind speed) across different burn days (Davies et al. 2010). In general this may mean
32 that managed fires, typically burnt during low-severity conditions, have more limited effects
33 than wildfires patterns are not consistent as wildfires can occur in a wide variety of conditions

1 and not all have particularly severe effects. Rather little effort has been made to try to capture
2 or understand the effects of such variation but this is vital in order to monitor the amount of
3 carbon released during wildfires and the extent of the environmental change they cause.

4 This research was initiated following severe wildfires during the springs of 2011 and 2012.
5 We aimed to assess how burn severity varied within and between individual wildfires, and to
6 define what the implications of such variation might be for carbon emissions due to wildfire
7 and on-going development of fire danger rating systems such as the Met Office Fire Severity
8 Index (MOFSI; Kitchen et al. 2006). MOFSI is based on the Canadian Fire Weather Index
9 System (FWI System; Van Wagner 1987) and has been implemented in Wales and England in
10 order to provide a forecast of “exceptional” conditions when it becomes permissible to close
11 open-access land under the Countryside and Rights of Way Act 2000. To date there have been
12 limited efforts to examine the relationship between the FWI System and fire severity in the
13 UK. There is some evidence its moisture codes relate fairly well to ground fuels’ (Legg et al.
14 2007) and peat (Krivtsov et al. 2008) moisture content, and that it can do a tolerable job of
15 discriminating periods of increased wildfire risk (Legg et al. 2007). We aimed to investigate
16 the relationship between fire severity and all the sub codes and indices of the FWI System but
17 were particularly interested in its response to variation in the DMC (Duff Moisture Code), the
18 DC (Drought Code) and the BUI (Build-up Index). The DMC and DC are designed to relate,
19 respectively, to the moisture content of duff (partly decomposed litter) and compacted deeper
20 organic layers. Such fuel layers bear some resemblance to the moss/litter layers and peat
21 deposits found in British peatland ecosystems. The BUI integrates DMC and DC to provide
22 an overall indication of fuel availability. Our specific objectives were to: develop a simple
23 methodology to assess variation in burn severity post-hoc; assess the extent to which burn
24 severity and fuel consumption vary within and between wildfires; and to investigate links
25 between burning conditions (fuel type and fire weather) and variation in burn severity and
26 fuel consumption.

27 **2 Material and methods**

28 **2.1 Study sites**

29 Monitoring was completed on five different wildfires (Table 1) that burnt British peatlands
30 during the springs of 2011 and 2012. Sites were selected from information on fires provided
31 by land-managers, public and private land-owners, government agencies and Fire and Rescue

1 Services. We selected five sites that represented fires displaying moderate to high burn
2 severity and the North-South and West-East range of bioclimatic conditions of the British
3 uplands.

4 Pre-fire biotic and abiotic conditions varied both within and between our study sites (Table 1).
5 Most locations in England were broadly classified as mires on deep peat with vegetation
6 dominated by *Calluna* and *Eriophorum vaginatum* L. along with species such as *Vaccinium*
7 *myrtillus* L., *Deschampsia flexuosa* (L.) Trin. and *Trichophorum caespitosum* (L.) Hartm.
8 Vegetation was underlain by mats of pleurocarpous mosses. A number of plot locations were
9 recorded at noticeably wetter locations. Here *Calluna* was less dominant, *Eriophorum* spp.
10 and *T. caespitosum* occasionally very abundant and ground layer vegetation included patches
11 of *Sphagnum*. Sites in Scotland represented opposite ends of the spectrum of peatland habitat
12 types found in Britain. Finzean was comparatively drier, had shallow, stony organic soils and
13 vegetation dominated by a mixture of *Calluna* and *Pteridium aquilinum* (L.) Kuhn. The site at
14 Loch Doon was a bog with true peat soils and vegetation dominated by *Molinia caerulea* (L.)
15 Moench, *Myrica gale* L. and *Sphagnum* spp.

16 **2.2 Field data collection**

17 Burn severity and fuel consumption sampling was performed approximately 6 months after
18 the fires occurred. Previous researchers have collected such data as much as a year after fire
19 (e.g. de Groot et al. 2009). Wildfires are sporadic, unpredictable events meaning sites had not
20 been surveyed prior to the burns. Similarly to other studies (e.g. Kasischke and Johnstone
21 2005, Hollis et al. 2007, de Groot et al 2009), we used paired plots with burnt/unburnt
22 subplots located across the fire perimeter (see Supplementary Material Figure 1). Two or
23 three paired plots were located within each fire and chosen to represent the range of burn
24 severities visible during a detailed site reconnaissance with local stakeholders. Many
25 peatlands in the British uplands have a patchwork of fuel structures produced by managed
26 burning. We were therefore careful to ensure that subplots were established where, following
27 observation of stem basal diameters, stem density and discussion with local land-managers,
28 we were confident that pre-fire fuel conditions across the fire-line were similar. Plots were
29 also only established in regions of the fireline known to have been actively extinguished. In
30 order to capture additional information about variation in burn severity we established a
31 number of unpaired plots within the interior of each fire (Table 1). Unburnt areas were not

1 available for comparison with these plots and they were only used to explore variation in burn
2 severity.

3 **2.3 Fire weather**

4 Variation in burning conditions between the fires was described using the FWI System (Van
5 Wagner 1987). The FWI System requires daily data on wind speed, temperature and humidity
6 at 12 noon as well as 24-hour accumulated rainfall. These were extracted from the British
7 Atmospheric Data Centre database for the nearest weather station to each of the wildfires
8 (mean distance = 15 km, max = 31 km). Rainfall data were available from rain gauges closer
9 to the fire site than the nearest full weather station and these to estimate precipitation (mean
10 distance = 5 km, max = 10 km). Available data on 24 hour accumulated rainfall was 09:00-
11 09:00 rather than noon to noon though the difference is unlikely to be of importance. FWI
12 System values were calculated using the package “fume” (Santander Meteorology Group
13 2012) in R 3.1.2 (R Development Core Team 2014). Some of the moisture codes and indices
14 of the FWI System have long lag times (52 days for the Drought Code) so values were
15 calculated with a 90 day lead-in.

16 **2.4 Assessing burn severity**

17 To assess burn severity we adapted the Composite Burn Index (Key and Benson 2006) which
18 was developed in the USA to allow semi-quantitative assessment of burn severity and ground-
19 truthing of remotely sensed data (e.g. Miller and Thode 2007). The CBI uses a scoring system
20 to visually estimate a fire’s impact on components of each five fuel strata. For instance,
21 assessment of “substrates” considers consumption of downed fuels of a variety of size classes
22 (litter up to heavy fuels > 8 inches diameter), consumption of duff layers and changes to the
23 cover and colour of soil and rock. Similarly to Schepers et al. (2014), we adapted the CBI to
24 account for the unique vertical structure and fuelbeds of treeless peatland habitats and,
25 specifically, to include the impact of fire on peat-building *Sphagnum* species. We recorded
26 severity in circular plots 20 m in diameter (Supplementary Material, Figure S1) according to
27 two strata – substrates (soil, litter and mosses) and the field layer (dwarf shrubs and
28 graminoids; see Supplementary Material, Table S1). All variables were rated on a scale of 1-3
29 with individual ratings averaged within strata and then summed across the strata. Any variable
30 that was not relevant, or which could not be recorded, for a particular plot was disregarded. A

1 full protocol and data collection sheet for using the peatland CBI methodology (pCBI) are
2 provided in the Supplementary Material.

3 **2.5 Estimating fuel consumption**

4 We assessed fuel consumption in two pCBI burnt-unburnt paired plots for each fire. Within
5 both the burnt and unburnt subplots we randomly located two fuel quadrats (0.25 m²) and five
6 gas-flux chambers (0.12 m²). All biomass above the top of the peat was harvested in each
7 quadrat/chamber. A total of fourteen biomass estimates were thus available for each plot -
8 seven from burnt and seven from unburnt subplots. Harvested vegetation was separated into
9 the following categories: dwarf shrubs, graminoids, ferns (*P. aquilinum*), pleurocarpous
10 mosses and plant litter, *Sphagnum* spp., tussock bases of *M. caerulea* and/or *Eriophorum* spp.
11 and woody stems buried in the moss and litter. During analysis, the first three categories were
12 grouped into a surface fuel category whilst the mosses, litter, tussock bases and buried stems
13 were classified as ground fuels. Material was dried for 48 h at 80 °C.

14 Fuel consumption in our wildfires was compared with values reported by Legg et al. (2007)
15 for 26 experimental prescribed burns in *Calluna*-dominated moorland fuel types. Legg et al.
16 (2007) used a non-destructive method, based on visual obstruction of a measuring stick
17 (Davies et al. 2008b), to estimate pre-fire surface and ground fuel loads. Post-fire surface fuel
18 loads were estimated via destructive harvesting. We estimated ground fuel consumption in
19 these fires by using the reported mean change in moss/litter layer depth following burning and
20 in the equation in Davies et al. (2008) which relates moss/litter layer depth D_m , (cm) to ground
21 fuel biomass (B_g , g m⁻²; equation 1).

$$22 \quad B_g = 407 + 171 \times D_m \quad (1)$$

23 **2.6 Data analysis**

24 We analysed burn severity data at the plot-level, in essence treating each plot as a separate
25 observation of fire effects and burn severity. We believe that this is valid because substantial
26 variations in vegetation type and fuel structure across the fire ground and changes in fire
27 weather during the course of the burn day mean fire behaviour can be considered independent
28 at each plot. This approach is frequently used in wildland fire research as obtaining numerous
29 observations of individual fires is often impossible (e.g. Fernandes et al. 2000, de Groot et al.
30 2009). The relationship between pCBI and FWI system codes was analysed graphically and

1 using correlation analysis (Pearson product-moment correlation in the “cor.test” function in R
2 3.1.2; R Core Team 2014).

3 We used a Generalised Linear Mixed Model (GLMM) with a normal error distribution to
4 investigate spatial variation in estimated fuel consumption. The aim of our analysis was to
5 partition variance in our data to understand how fuel consumption varies at multiple scales
6 (i.e. between fires, plots within fires and within plots) and how this contributes to uncertainty
7 in estimates of fuel consumption. We were not interested in testing the hypothesis that there is
8 a difference in biomass between burnt and unburnt plots as this is not particularly
9 enlightening. The GLMM was run with plot and fire site were defined as random effects
10 whilst status (burnt/unburnt) and sample type (chamber/quadrat) were defined as fixed effects.
11 Including plot as a random effect accounts for the paired burnt-unburnt subplots design of our
12 experiment. We selected the best fitting model by comparing a full model and a minimal
13 model. The minimal model contained all sources of variation intrinsic to the design: the main
14 effects of status and sample type, and random intercepts at the plot and fire levels. The full
15 model additionally allowed the effect of status to vary between sample types (fitted as an
16 interaction between status and sample type), and between plots and fires (fitted as random
17 slopes at the plot and fire levels). Analysis started with the full model and simplification
18 proceeded by null hypothesis testing, dropping non-significant effects. Random effects were
19 tested first, using parametric bootstrapping with 10,000 replicates (Faraway 2005), dropping
20 effects where $P > 0.1$. Fixed effects were then tested using likelihood ratio tests, dropping
21 effects where $P > 0.05$. We justify using a less stringent significance level for random effects
22 on the basis that power for testing random effects is generally low with few random effect
23 levels, and incorrectly dropping a random effect due to a false negative test result can result in
24 over-precise (anti-conservative) fixed effect estimates (Schielzeth and Forstmeier 2009). We
25 used parametric bootstrapping with 10,000 replicates to estimate confidence intervals around
26 mean plot-level consumption. This process was used to fit separate models for both ground
27 and surface fuel consumption. Fuel consumption was square-root transformed to improve the
28 fit of the residuals to a normal distribution. Log transformation was also considered but
29 provided a poorer fit (see Appendix II for dot plots showing the raw data distributions,
30 Supplementary Material). There is debate in ecology about the usefulness of P-values (Ellison
31 et al. 2014) and we do not report them. Rather we report the explanatory power of the final
32 selected models and the variance explained by different levels of our experimental design
33 Thus, for the final, reduced models we used the procedures described by Nakagawa &

1 Schielzeth (2013) and Johnson (2014) to calculate marginal and conditional R² values. These
2 describe the explanatory power of the fixed effects and the whole model (fixed + random
3 effects) respectively. As an initial step in this analysis we were also able to partition the
4 variance in our data into that related to the fixed effects and the random effects of plot and
5 fire. We assumed that residual variance was the result of within subplot variation in load
6 between samples.

7 We examined controls on mean fuel consumption by combining the estimates of fuel
8 consumption during wildfires produced by the GLMM analysis with information available
9 from the prescribed fires reported by Legg et al. (2007). This allowed us to examine how
10 mean ground and surface fuel consumption varies over a wider range of fire weather
11 conditions. We used the “lm” function in R to model changes in the consumption and
12 combustion completeness of surface and ground fuels as a function of pre-fire fuel load and
13 fire weather.

14 **3 Results**

15 **3.1 Variation in burn severity**

16 There was substantial variation in burn severity both within and between individual fires
17 (Figure 1). On average, mean pCBI varied 1.6 fold between wildfires but up to 1.7 fold within
18 fires. Variability in burn severity was particularly substantial in the Anglezarke and Loch
19 Doon wildfires. Examining the relationship between plot vegetation community, fire weather
20 conditions and pCBI suggested potential interactions between these variables (Figure 2). In
21 general, pCBI appeared to increase with higher DMC ($r = 0.80$, $P = 6.4 \times 10^{-4}$) and DC ($r =$
22 0.68 , $P = 7.9 \times 10^{-3}$) values. Plots in drier *Calluna*-dominated communities (National
23 Vegetation Community H12) appeared to burn at high severities at lower DMC and DC
24 values than wetter bog and mire communities (NVC M19, M20, M25a). However, fire sites
25 with more varied vegetation community structure did not necessarily show the greatest
26 amount of variation in fire severity.

27 **3.2 Variation in fuel consumption**

28 Both surface and ground fuel consumption were best represented by a model which included
29 the fixed effects of plot status (burnt/unburnt), sample type (quadrat/chamber), random
30 intercepts for individual fires and plots within fires, and random slopes for the effect of status

1 within individual plots (Table 2). Plot status had considerably greater explanatory power for
2 surface fuel loads compared to ground fuel loads where random factors attributable to
3 variation in load between fires, plots and samples explained a greater proportion of the
4 variance. There was considerable variation in fuel consumption both within and between
5 different wildfires (Figure 3), indeed variability within some fires was greater than that seen,
6 on average, between fires.

7 For surface fuels there was a positive linear relationship between pre-fire fuel load and mean
8 fuel consumption irrespective of fire type (Figure 4). Surface fuel consumption (C_s) was best
9 predicted by pre-fire fuel load (L_s ; $R^2_{adj} = 0.73$, $P = 1.79 \times 10^{-11}$; Equation 2). None of the
10 FWI System values were significant or substantially improved the model fit. For ground fuels,
11 the relationship between pre-fire fuel load and mean fuel consumption was noticeably
12 different with a positive, linear relationship for wildfires but little change in consumption with
13 load for prescribed fires (Figure 4). Ground fuel consumption and ground fuel combustion
14 completeness appeared to decline with ground fuel load (Figure 4). It proved difficult to
15 develop a satisfactory model of ground fuel consumption, but ground fuel combustion
16 completeness (P_g) could be predicted tolerably well as an asymptotic function of BUI (B ; R^2_{adj}
17 $= 0.77$, $P = 1.77 \times 10^{-12}$; Equation 3).

$$18 \quad C_s = 0.173 + 0.624 \times L_s. \quad (2)$$

$$19 \quad P_g = \sqrt{-0.034 + 0.020 \times B}, \quad (3)$$

20 **4 Discussion**

21 Wildfires are variable in every aspect and the fires we were able to assess do not capture the
22 full range of possible conditions. Notably, none of our fires displayed peat smouldering
23 outside of isolated “hotspots”. Nevertheless, this work represents the first multi-site attempt to
24 investigate the relationship between burning conditions and wildfires’ ecosystem effects on
25 moorlands. Wildfires on peatlands are recognised as a growing global challenge with the
26 potential to develop into a significant positive feedback to climate change (Kettridge et al.
27 2015). Scientists and land-managers currently have limited understanding of the extent and
28 causes of variation in the severity and ecological effects of temperate peatland fires.
29 Temperate peatlands, such as those found in the UK, are likely to be at the forefront of the
30 effects of climate change with some studies suggesting considerable declines in their
31 bioclimatic space (Gallego-Sala and Prentice 2013) and fundamental changes in state

1 associated with even moderate reductions in water tables (Kettridge et al. 2015). UK
2 peatlands are of particular management concern due to the substantial area that has already
3 been lost or degraded by changing land-management, and debates over the effects of
4 traditional managed burning on the ecosystem services they provide (Bonn et al. 2009).

5 The growing peatland wildfire problem demands evidence to inform management and solve
6 on-going conflict about the impacts of burning. Our adapted version of the CBI provides a
7 method for rapid cataloguing of post-fire effects and burn severity in UK peatland
8 ecosystems. The pCBI method appeared to function well and detected substantial differences
9 in burn severity between and within individual fires. Importantly, there was evidence that
10 increased pCBI can be attributed to reduced ground fuel layer moisture content as higher burn
11 severity was recorded at higher values of DMC and DC (Figure 2). Our results, and existing
12 evidence that the DC may relate to the potential for smouldering peat fires (Davies et al.
13 2013), raise the prospect that it will be possible to forecast the potential for damaging
14 wildfires. There was also a suggestion that burn severity is a function of ecosystem type, and
15 associated site hydrology, as we recorded higher severity burns in dry moorland sites than
16 would be expected given the intermediate DMC/DC values at which they occurred (Figure 2).
17 Sites with thin organic soils may thus be at greater risk of severe and smouldering wildfires
18 than those with deeper peat and forecasts for such systems should be developed separately.

19 **4.1 Variation in fuel consumption**

20 In general, the random factors in our GLMM, that account for variation in loads within plots
21 and within and between fires, explained for as much, if not more, of the variation in fuel loads
22 across our survey than differences between burnt/unburnt subplots. This was particularly true
23 for ground fuels where 70% of the variance was attributable to spatial variation rather than the
24 effects of fire or sample type. For ground fuels, the higher variance explained by random
25 factors is possibly a function of the substantial differences in their composition between sites
26 and, at some locations, between plots. Our sites included both bog communities with
27 substantial cover of *Sphagnum* spp. and drier sites with thin organic soils where bryophyte
28 communities were poorly-developed and ground fuels were dominated by litter. When
29 considering the wildfires alone, ground fuel consumption showed a linear relationship with
30 pre-fire fuel load though there was some evidence of a possible interaction with ecosystem
31 type. Shetler et al. (2008) demonstrated that the presence of *Sphagnum* had a limiting effect
32 on total carbon release during fires in black spruce forest peatlands and combustion

1 completeness was lowest at Loch Doon, the wettest of our sites, where *Sphagnum* spp. and
2 *Molinia* tussocks comprised a substantial proportion of the ground fuel load (Figure 4).
3 However, this fire also occurred under the least severe fire weather conditions (Figure 2).

4 When we analysed ground fuel consumption for the wild and prescribed fires together we
5 were unable to develop a tolerably robust model. We hypothesise that this was due to
6 differences in combustion rates between ecosystem types. Given that all our prescribed fires
7 were in drier *Calluna*-dominated heathlands, our current data set was not sufficient to model
8 ecosystem-specific rates. de Groot et al. (2009) examined variation in ground fuel
9 consumption, albeit in non-peatland systems, and also found differences in the controlling
10 relationships for different fuel types. We were, however, able to predict combustion
11 completeness based on BUI. These results are significant because: i) it provides further
12 evidence that the moisture status of ground fuel layers is a critical control on burn severity in
13 peatlands; ii) it further demonstrates that certain components of the FWI System (DMC, DC
14 and BUI) may be useful in forecasting potential burn severity.

15 Surface fuel consumption also showed significant spatial variation, though variability
16 between plots explained a greater proportion of the variance than that between fires (Table 2).
17 Surface fuel consumption of shrubs and graminoids was strongly related to pre-fire fuel load
18 (Figure 4, Equation 2) and there was no significant effect of fire weather conditions. This
19 matches some of our existing understanding of fire behaviour in moorland fuel types (Legg et
20 al. 2007). In the vast majority of cases a relatively constant proportion of fuel is consumed as
21 the fire spreads through the *Calluna* canopy consuming fine fuel particles but leaving larger
22 live basal stems unburnt. Coarser fuels form a larger proportion of the fuel in older stands
23 (Davies et al. 2008b) but rarely burn except under exceptionally severe conditions. This
24 accounts for the decline in combustion completeness with increasing fuel load (Figure 4). The
25 variability we recorded in fuel consumption within and between our fires is likely to be
26 attributable to i) differences in fuel load between ecosystems; and ii) the highly-managed
27 nature of many UK peatlands where rotational patch burning produces a mosaic of
28 fuel/habitat loads across the landscape

29 Assuming that the approximate carbon content of our fuels was 49% (Worrall et al. 2013), our
30 data suggests that average carbon release from the combustion of above-ground biomass by
31 wildfires can range between 0.36 kg C m⁻² and 1.00 kg C m⁻². This is a somewhat greater than
32 seen for the prescribed fires which saw C release rates of between 0.26 kg C m⁻² and 0.66 kg

1 C m⁻². Our wildfire C release rates are considerably higher than the mean release of 0.15 kg C
2 m⁻² reported by Clay and Worrall (2011) for the single moorland wildfire they studied, but
3 both their result and ours falls within the range reported by Poulter et al. (2006) for a
4 temperate peatland wildfire in North Carolina, USA. Whether or not leaving a peatland
5 unburnt would increase the amount of carbon stored in the landscape is difficult to judge from
6 our data alone. Whilst unmanaged peatlands may store greater amounts of C in surface and
7 ground fuel layers than those that are subject to regular managed burning, they may also be
8 more susceptible to large-scale wildfires because of their unmanaged fuel loads (Allen et al.
9 2013). Our results show rates of fuel consumption during such wildfire events will also be
10 higher.

11 **5 Conclusions**

12 Burn severity varies considerably in relation to fuel structure and fire weather. To date much
13 of the research on the effects of fire on moorlands has drawn an artificial distinction between
14 the effects of prescribed burning and wildfires, though the latter do seem to be associated with
15 increased severity. Our results suggest that critical differences in burn severity and fuel
16 consumption can be linked to the flammability of ground fuel layers. Our data add to the
17 information available to researchers modelling the effects of land-management and fire
18 regimes on ecosystem carbon dynamics but we urge caution in their use and suggest that
19 further work to determine linkages between burning conditions and both short- and-long term
20 fire effects is urgently needed in temperate peatland ecosystems.

21 **6 Author contributions**

22 GMD designed the study, assisted with fieldwork, analysed the data and wrote the
23 manuscript; RD completed the fieldwork and assisted in analysing the data and writing the
24 manuscript; PJ designed the mixed effects modelling element of the analysis and wrote the
25 paper; AG designed the study and helped write the paper.

26 **7 Acknowledgements**

27 Funding for this research was provided by the National Environment Research Council
28 (project number NE/J006289/1). Additional support was kindly provided by Jane and David
29 Philbrick. A large number of individuals and organisations provided us with information on
30 wildfire events but we are particularly grateful to the England and Wales Wildfire Forum, the
31 Knowledge for Wildfire project at the University of Manchester and Michael Bruce. Access

1 to field study sites and logistical support were provided by United Utilities, Yorkshire Water,
2 the National Trust, Finzean Estate, and Forestry Commission Scotland. Field data collection
3 was completed with the assistance of Sophie Philbrick. Two anonymous reviewers helped us
4 improve our manuscript
5

1 **References**

- 2 Allen, K. A., Harris, M. P. K., and Marrs, R.H.: Matrix modelling of prescribed burning in
3 *Calluna vulgaris*-dominated moorland: short burning rotations minimize carbon loss at
4 increased wildfire frequencies. *J. Appl. Ecol.*, 50, 614-624, 2013.
- 5 Bates, D., Maechler, M., Bolker, B. and Walker, S.: lme4: Linear mixed-effects models using
6 Eigen and S4. R package version 1.1-7, <http://CRAN.R-project.org/package=lme4>, 2014.
- 7 Bonn, A., Allott, T., Hubacek, K. and Stewart, J.: Drivers of Environmental Change in the
8 Uplands, Routledge, London, U.K, 2009.
- 9 Bradley, R. I., Milne, R., Bell, J., Lilly, A., Jordan, C. and Higgins, A.: A soil carbon and land
10 use database for the United Kingdom. *Soil Use Manage.*, 21, 363-369, 2005.
- 11 Clay, G. D., and Worrall, F.: Charcoal production in a UK moorland wildfire – how important
12 is it? *J. Environ. Manage.*, 92, 676-682, 2011.
- 13 Davies, G.M., Gray, A., Hamilton, A. and Legg, C.J.: The future of fire management in the
14 British uplands. *Int. J. Biodiv. Sci. Manage.*, 4, 127-147, 2008a
- 15 Davies, G. M., Legg, C. J., Hamilton, A., and Smith, A. A.: Using visual obstruction to
16 estimate heathland fuel load and structure, *Int. J. Wildland Fire*, 17, 380-389, 2008b.
- 17 Davies, G. M., Smith, A. A., McDonald, A. J., Bakker, J. D., and Legg, C. J.: Fire intensity,
18 fire severity and ecosystem response in heathlands: factors affecting the regeneration of
19 *Calluna vulgaris*, *J. Appl. Ecol.*, 47, 356-365, 2010.
- 20 Davies, G. M., Gray, A., Rein, G., and Legg, C. J.: Peat consumption and carbon loss due to
21 smouldering wildfire in a temperate peatland. *For. Ecol. Manage.* 308, 136-144, 2013.
- 22 de Groot, W. J., Pritchard, J. M., and Lynham, T. J.: Forest floor fuel consumption and carbon
23 emissions in Canadian boreal forest fires. *Can. J. For. Res.*, 39, 367-382, 2009.
- 24 Ellison, A. M., Gotelli, N. J., Inouye, B. D., and Strong, D. R.: P values, hypothesis testing,
25 and model selection: it's déjà vu all over again, *Ecology*, 95, 609-610, 2014.
- 26 Faraway, J. J.: Extending the Linear Model with R: Generalized Linear, Mixed Effects and
27 Nonparametric Regression Models. Chapman and Hall/CRC, Boca Raton, FL, USA, 2005.
- 28 Fernandes, P. M., Catchpole, W. R. and Rego, F. C. Shrubland fire behaviour modelling with
29 microplot data. *Can. J. For. Res.*, 30, 889-899, 2000.

1 Gallego-Sala, A.V., and Prentice, I. C.: Blanket peat biome endangered by climate change.
2 Nat. Clim. Change, 3, 152-155, 2012.

3 Grant, M. C., Mallord, J., Stephen, L., and Thompson, P.S.: The costs and benefits of grouse
4 moor management to biodiversity and aspects of the wider environment: a review. RSPB
5 Research Report Number 43. RSPB, Sandy Bedfordshire, UK.
6 http://www.rspb.org.uk/Images/grant_mallord_stephen_thompson_2012_tcm9-318973.pdf,
7 2012.

8 Hogg, P., Squires, P. and Fitter A.H.: Acidification, nitrogen deposition and rapid
9 vegetational change in a small valley mire in Yorkshire. Biol. Conserv., 71, 143-153, 1995.

10 Holden, J., Chapman, P. J. and Labadz J. C.: Artificial drainage of peatlands: hydrological
11 and hydrochemical process and wetland restoration. Prog. Phys. Geog., 28, 95-123, 2004.

12 Hollis, J.J., Matthews, S., Anderson, W.R., Cruz, M.G., and Burrows, N.D.: Behind the
13 flaming zone: Predicting woody fuel consumption in eucalypt forest fires in southern
14 Australia. For. Ecol. Manag., 261, 2049-2067, 2011.

15 Johnson, P.C.D.: Extension of Nakagawa and Schielzeth's R^2 GLMM to random slopes
16 models. Methods Ecol. Evol., 5, 944-946, 2014.

17 Kaat, A. and Joosten, H.: Factbook for UNFCCC policies on peat carbon emissions. Wetlands
18 International, Wageningen, The Netherlands,
19 <http://www.wetlands.org/Portals/0/publications/Report/fact%20book%20for%20unfcc%20pol>
20 [icies%20on%20peat%20carbon%20emissions%20for%20web.pdf](http://www.wetlands.org/Portals/0/publications/Report/fact%20book%20for%20unfcc%20policies%20on%20peat%20carbon%20emissions%20for%20web.pdf), 2009.

21 Kasischke, E.S., and Johnstone, J. F.: Variation in postfire organic layer thickness in a black
22 spruce forest complex in interior Alaska and its effects on soil temperature and moisture. Can.
23 J. For. Res., 35, 2164-2177, 2005.

24 Kettridge, N., Turetsky, M. R., Sherwood, J. H., Thompson, D. K., Miller, C. A., Benscoter,
25 B. W., Flannigan, M. D., Wotton, B. M., and Waddington, J. M.: Moderate drop in water
26 table increases peatland vulnerability to post-fire regime shift. Sci. Rep., 5, 8063, 2015.

27 Key, C.H. and Benson, N. C.: Landscape Assessment LA Sampling and Analysis Methods.
28 General Technical Report RMRS-GTR-164-C. USDA Forest Service Rocky Mountain
29 Research Station, Fort Collins, CO, USA.
30 http://www.fs.fed.us/rm/pubs/rmrs_gtr164/rmrs_gtr164_13_land_assess.pdf, 2006.

1 Kitchen, K., Marno, P., Legg, C. J., Bruce, M., and Davies, G M.: Developing a fire danger
2 rating system for the United Kingdom. *For. Ecol. Manag.*, 234S1, S21, 2006.

3 Krivtsov, V., Gray, A., Valor, T., Legg, C. J., and Davies, G. M.: The fuel moisture content of
4 peat as a fuel in relation to meteorological factors, *WIT Trans. Ecol. Env.*, 119, 193-200,
5 2008.

6 Legg, C.J., Davies, G. M., Marno, P., and Kitchen, K.: Developing a Fire Danger Rating
7 System for the UK: FireBeaters final report. Report to the Scottish Wildfire Forum.
8 <https://www.era.lib.ed.ac.uk/handle/1842/3011>, 2007.

9 Maltby, E., Legg, C. J., and Proctor, M. C. F.: The ecology of severe moorland fire on the
10 North York Moors: Effects of the 1976 fires, and subsequent surface and vegetation
11 development, *J. Ecol.*, 78, 490-518, 1990.

12 Miller, J. D., and Thode, A. E.: Quantifying burn severity in a heterogeneous landscape with a
13 relative version of the delta Normalized Burn Ratio dNBR, *Remote Sens. Environ.*, 109, 66-
14 80, 2007.

15 Mitra, S., Wassmann, R., and Vlek, P. L. G.: An appraisal of global wetland area and its
16 organic carbon stock, *Curr. Sci.*, 88, 25-35, 2005.

17 Moore, P.D.: The future of cool temperate bogs. *Environ. Conserv.*, 29, 3-20, 2002.

18 Nakagawa, S. and Schielzeth, H.: A general and simple method for obtaining R^2 from
19 generalized linear mixed-effects models, *Methods Ecol. Evol.*, 4, 133-142, 2013.

20 Ostle, N. J., Levy, P. E., Evans, C. D., and Smith, P.: UK land use and soil carbon
21 sequestration. *Land Use Policy*, 26S1, S274-S283, 2009.

22 Poulter, B., Christensen, N. L. Jr., and Halpin, P. N.: Carbon emissions from a temperate peat
23 fire and its relevance to interannual variability of trace atmospheric greenhouse gases. *J.*
24 *Geophys. Res.: Atmos.*, 111, D6, 2006.

25 R Development Core Team: R: A language & environment for statistical computing. R
26 Foundation for Statistical Computing, Vienna, Austria. <http://www.R-project.org>, 2014.

27 Rodwell, J.S.: *British Plant Communities. Volume 2. Mires & heaths.* Cambridge University
28 Press, Cambridge, UK, 1991.

- 1 Santander Meteorology Group: fume: FUME package. R package version 1.0.
2 <http://CRAN.R-project.org/package=fume>, 2012.
- 3 Schepers, L., Haest, B., Veraverbeke, S., Spanhove, T., Vanden Borre, J., and Goossens, R.:
4 Burned area detection and burn severity assessment of a heathland fire in Belgium using
5 airborne imaging spectroscopy (APEX). *Remote Sens.*, 6, 1803-1826, 2014.
- 6 Schielzeth, H. and Forstmeier, W.: Conclusions beyond support: overconfident estimates in
7 mixed models. *Behav. Ecol.*, 20, 416-420, 2009.
- 8 Scottish Government: The Muirburn Code. Scottish Government, Edinburgh, 2011.
- 9 Shetler, G., Turetsky, M. R., Kane, E., and Kasischke, E.: *Sphagnum* mosses limit total
10 carbon consumption during fire in Alaskan black spruce forests, *Can. J. For. Res.*, 38, 2328-
11 2336, 2008.
- 12 Stace, C. *New Flora of the British Isles*. Cambridge University Press, Cambridge, U.K, 1997.
- 13 Turetsky, M. R., Benscoter, B., Page, S., Rein, G., van der Werf, G. R., and Watts, A.: Global
14 vulnerability of peatlands to fire and carbon loss, *Nat. Geosci.*, 8, 11-14, 2015.
- 15 Van Wagner, C.E.: Development and structure of the Canadian Forest Fire Weather Index
16 System. Forestry Technical Report 35. Canadian Forestry Service, Ottawa, 1987.
- 17 Watts, A. C., and Kobziar, L. N.: Smoldering combustion and ground fires: ecological effects
18 and multi-scale significance. *Fire Ecol.*, 9, 124-132, 2013.
- 19 Worrall, F., Rowson, J. G., Evans, M. G., Pawson, R., Daniels, S., and Bonn, A.: Carbon
20 fluxes from eroding peatlands – the carbon benefit of revegetation following wildfire, *Earth*
21 *Surf. Process. Landf.*, 36, 1487-1498, 2011.
- 22 Worrall, F., Clay, G. D., and May, R.: Controls upon biomass losses and char production from
23 prescribed burning on UK moorland, *J. Environ. Manage.*, 120, 27-36, 2013.
- 24 Yu, Z., Loisel, J., Brosseau, D. P., Beilman, D. W., and Hunt, S. J.: Global peatland dynamics
25 since the Last Glacial Maximum, *Geophys. Res. Lett.*, 110, L13402, 2010.

1 **Table 1:** Summary of wildfires used in this study including variation in biotic and abiotic conditions across the fire grounds and the monitoring effort associated with each
 2 fire. Vegetation type is reported as National Vegetation Classification (NVC) communities (Rodwell 1991) with the NVC code given in brackets. Paired CBI plots were
 3 those placed around the fire perimeter to enable direct comparison of burnt and unburnt fuel loads and soil gas fluxes. Stand-alone CBI plots were additional plots located
 4 within the fire in order to give a more comprehensive overview of variation in burn severity. Nearby unburnt comparison locations were not available for these plots

Fire name and Location	Latitude and Longitude	Date of fire	Burned area (ha)	Elevation (m)	Soil	Vegetation type	Paired CBI plots	Stand-alone CBI plots
Anglezarke (N England)	53.658°N 2.569°W	29/Apr/2011	4,144	270 - 380	Deep peat	<i>Calluna vulgaris</i> - <i>Eriophorum vaginatum</i> blanket mire (M19)	3	4
Mardsen (N England)	53.596°N 1.976°W	09/Apr/2011	316	385 - 480	Deep peat	<i>Calluna vulgaris</i> - <i>Eriophorum vaginatum</i> blanket mire (M19) <i>Calluna vulgaris</i> - <i>Vaccinium myrtillus</i> heath (H12)	3	2
Loch Doon (SW Scotland)	55.214°N 4.393°W	29/May/2011	No data	230 - 250	Shallow peat	<i>Molinia caerulea</i> - <i>Potentilla erecta</i> mire (M25a)	2	2
Wainstalls (N England)	53.777°N 1.928°W	30/Apr/2011	82	385 - 420	Deep peat	<i>Calluna vulgaris</i> - <i>Eriophorum vaginatum</i> blanket mire (M19) Scattered <i>Calluna vulgaris</i> - <i>Vaccinium myrtillus</i> heath (H12)	3	3

Finzean	57.025°N	30/Mar/2012	19	320 - 340	Rocky	<i>Calluna vulgaris</i> - <i>Vaccinium</i>	3	0
(NE Scotland)	2.702°W				organic	<i>myrtillus</i> heath (H12)		

1

2

1 Table 2: Summary of the linear mixed model analyses on surface and ground fuel
 2 consumption showing, top - the proportion of variance explained by each component of the
 3 models (the marginal and conditional R^2 , respectively, show the explanatory power of the
 4 fixed effects and the whole model); and bottom – the magnitude of the fixed-effects’ terms in
 5 the model where “Estimate” is the increase in the square-root of fuel load in comparison to
 6 the reference level (Burnt or Gas flux chamber for status and sample type respectively).

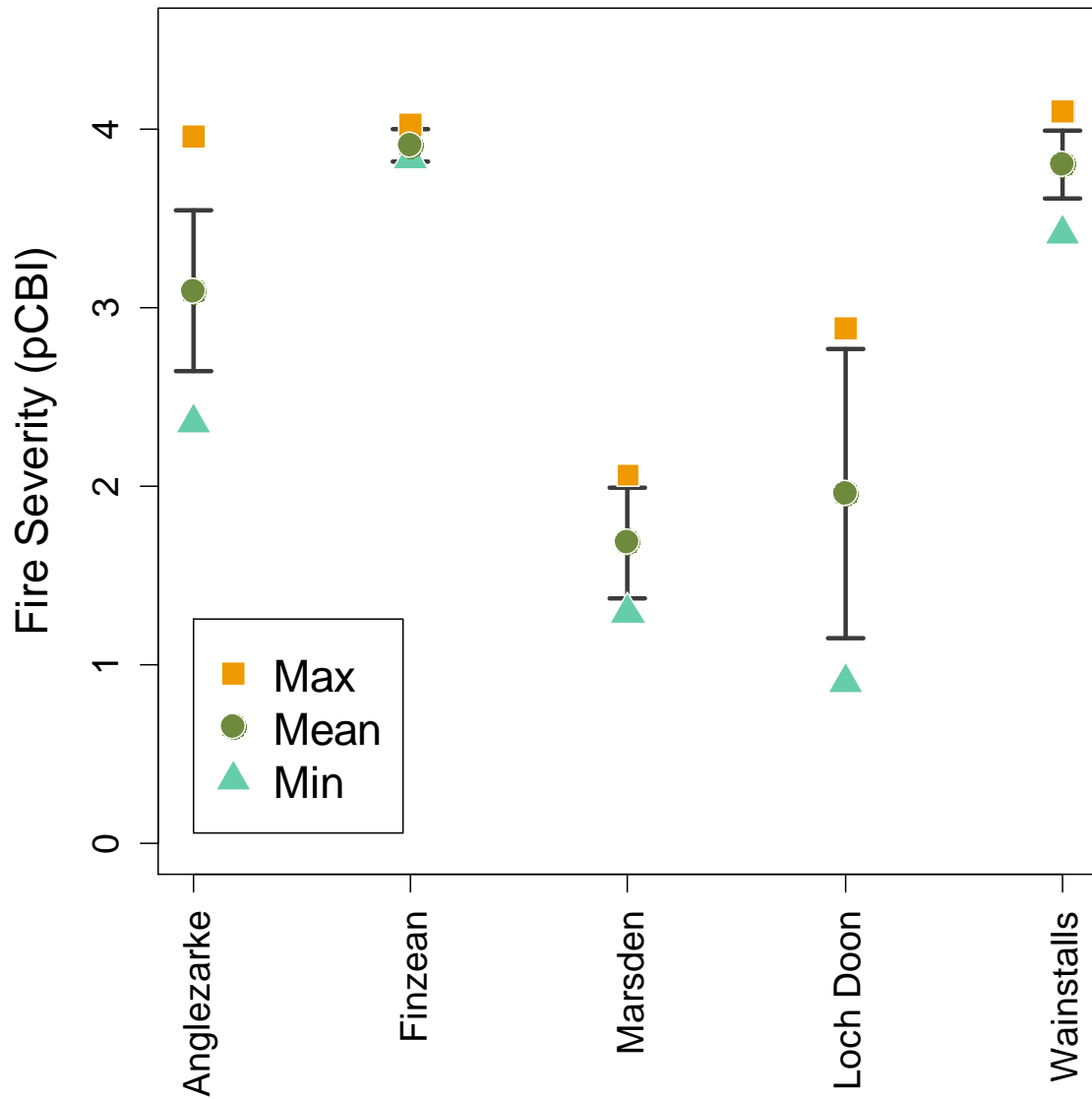
Model	Fixed effects (marginal R^2)	Random effects		Fixed + random effects (conditional R^2)	Residual
		Fire	Plot		
Surface fuels	48	5	24	77	23
Ground fuels	30	12	29	71	29

7

Model	Fixed effect	Estimate	S.E.	t
Surface fuels	Status - Unburnt	0.51	0.056	9.11
	Sample - Quadrat	0.11	0.034	3.27
Ground fuels	Status - Unburnt	0.49	0.074	6.62
	Sample - Quadrat	0.10	0.046	2.23

8

9

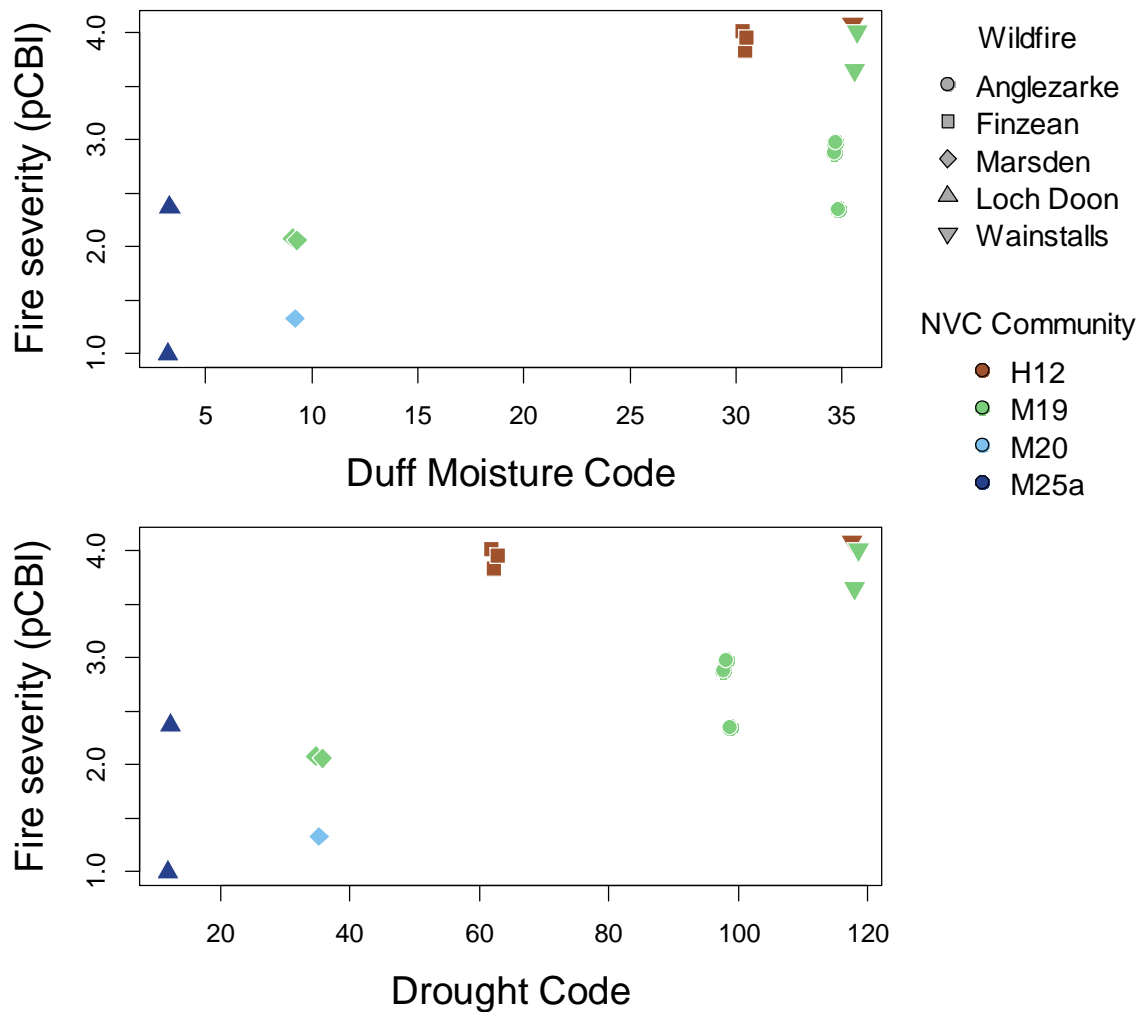


1

2 Figure 1: Variation in burn severity (peatland Composite Burn Index; pCBI) within and
 3 between five UK wildfires and across all 25 paired and unpaired pCBI plots. Error bars are
 4 95% confidence intervals for the mean.

5

1

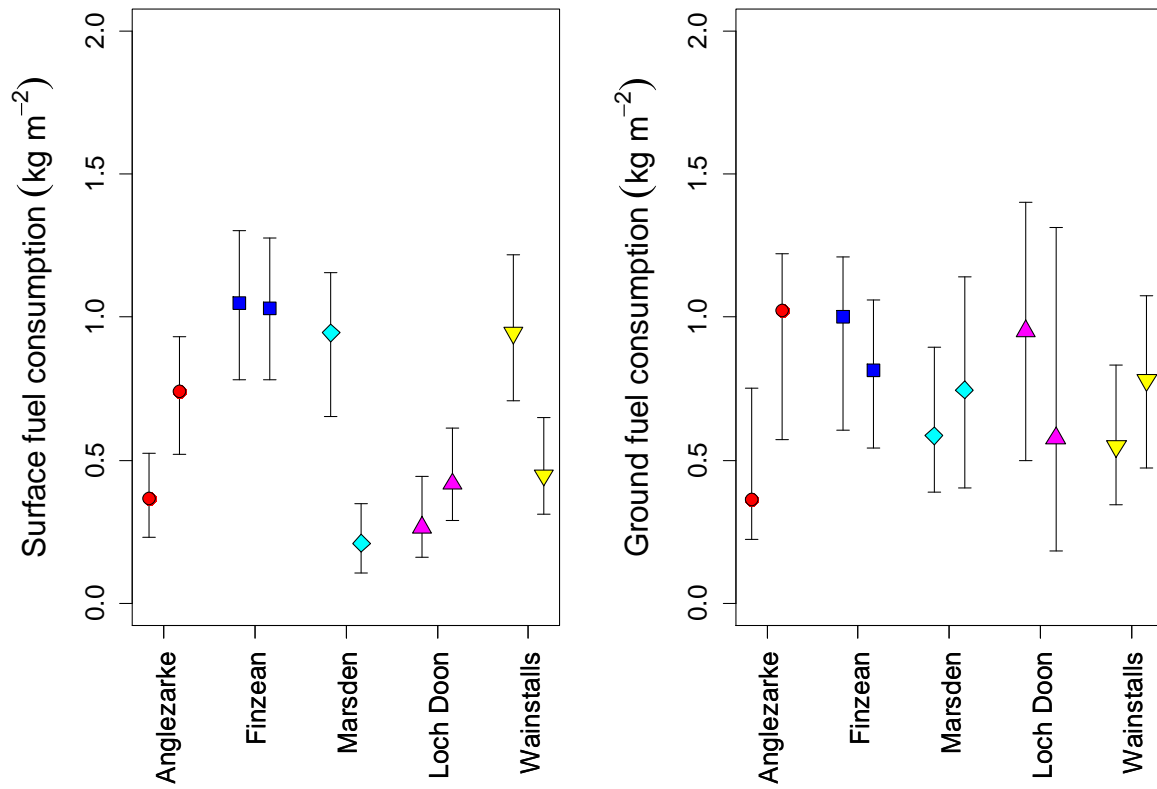


2

3 Figure 2: The relationship between burn severity as estimated by the peatland Composite
4 Burn Index, ecosystem type (National Vegetation Classification community; Rodwell 1991)
5 and moisture codes of the Canadian Fire Weather Index system. Only data for the 14 paired
6 burnt-unburnt pCBI plots were available. Codes shown are the Duff Moisture Code (DMC;
7 relating to loosely compacted organic layers of moderate depth) and the Drought Code (DC;
8 relating to the moisture content of peat and layers of organic soil). Individual wildfires are
9 shown as different symbol shapes, colours relate to NVC vegetation community (see Table 1
10 for NVC community descriptions).

11

1



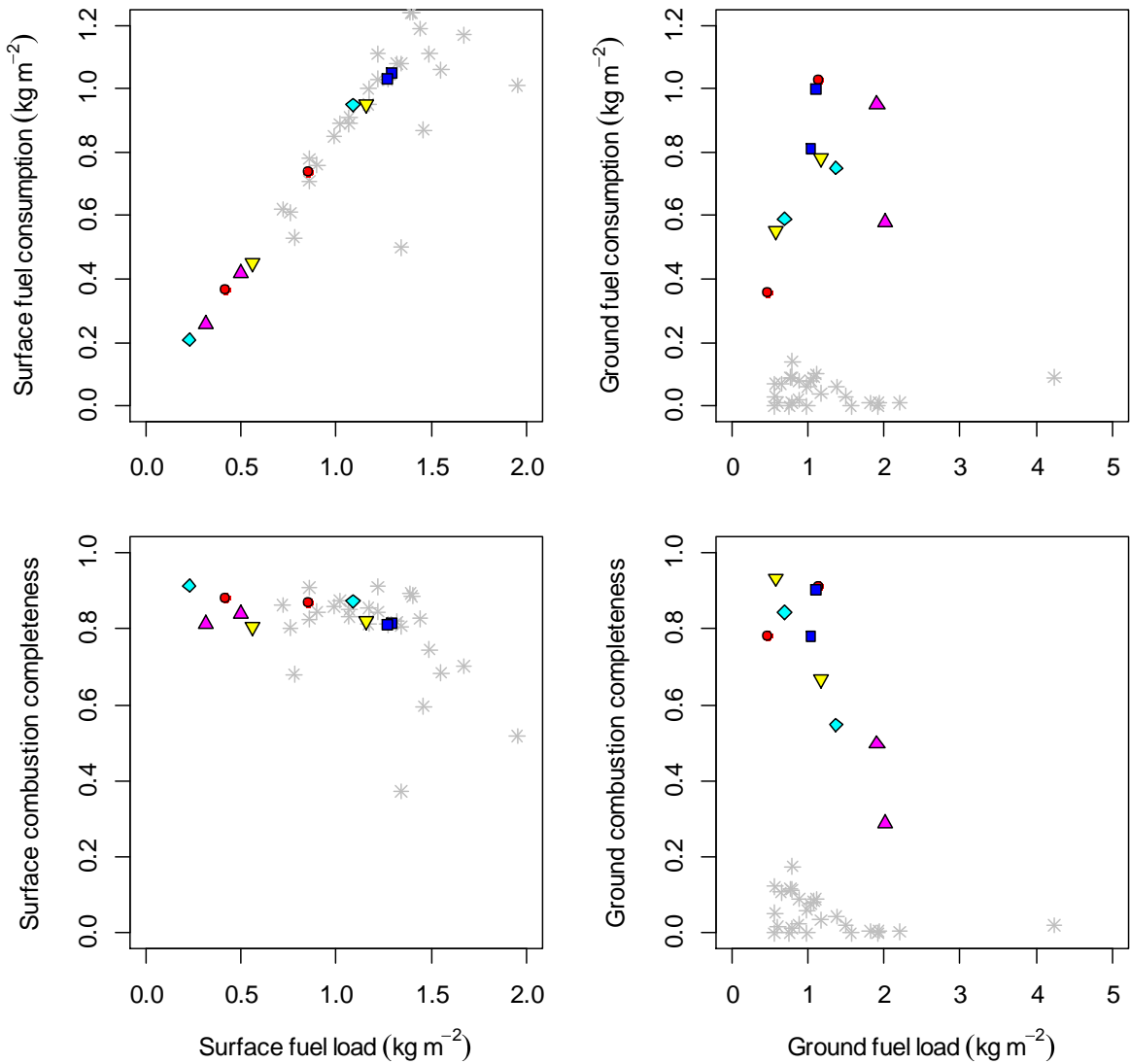
2

3 Figure 3: Estimated mean consumption of (left) surface, and (right) ground fuels across two
4 plots on each of five UK wildfires. Error bars are 95% confidence intervals estimated using
5 parametric bootstrapping based on a general linear mixed model analysis of variation in
6 consumption (Table 3).

7

8

1



2

3 Figure 4: The relationships between mean pre-fire fuel load and mean fuel consumption for
4 surface and ground fuels (top left and right respectively); and mean pre-fire fuel load and
5 mean combustion completeness of surface and ground fuels (bottom left and right). Stars are
6 experimental prescribed burns (see Legg et al. 2007), all other symbols are wildfires. The
7 colours and shapes of the points for wildfires follows Figure 3.

8